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ABSTRACT

Canonical correlation analysis is a parsimonious way of breaking down the association between two sets of variables through the use of linear combinations. As a result of the analysis, many types of coefficients can be generated and interpreted. These coefficients are only considered stable and reliable if the number of subjects per variable is sufficiently large. The first of these coefficients, the canonical correlation, is the bivariate correlation between the composite scores for the two sets of variables. Two additional coefficients, the canonical function and structure coefficients, address the contribution a single variable makes to the explanatory power of the set of variables to which the variable belongs. The communality coefficient explains how useful the variable is in defining the canonical solution. The adequacy coefficient indicates how adequately the analysis represents the total variance in the unweighted set. The extent to which a variable contributes to explaining the composite of the variable set to which the variable of interest does not belong is the index coefficient. A final outcome from canonical correlation analysis is the redundancy coefficient, which indicates the average proportion of variance for variables in one set that is reproducible with the variables in the other set. While the coefficient is easy to calculate, it is not recommended for interpretation in most cases. (Contains 3 tables and 10 references.) (SLD)

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# USING CANONICAL CORRELATION TO EXPLORE RELATIONSHIPS BETWEEN SETS OF VARIABLES: AN APPLIED EXAMPLE WITH INTERPRETIVE SUGGESTIONS

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## Introduction

Canonical correlation analysis is a procedure for exploring the relationship between two sets of variables containing two or more variables each. As argued by Baggaley (1981), the multivariate technique is the most general case of the general linear model. It is usually employed because the researcher wants to consider the simultaneous workings of all the variables of interest at once. As noted by Thompson (1984), canonical correlation analysis is appealing because although multivariate in nature, it can be presented in bivariate terms. This is achieved by the calculation of many different coefficients and correlations, each of which answer research questions detailing different aspects of the analysis. This paper presents seven different coefficients that are generated as a result of canonical correlation analysis. Each coefficient is described and interpreted using a practical example.

## Data Example

To illustrate a practical application of canonical correlation, data from a large West coast law school were used to determine the relationship between two sets of variables and their influence on admitted students' decisions whether or not to attend the law school. Eight variables were analyzed using SAS statistical software for the dataset containing 251 observations. The predictor set contained five variables that were measurements of the influence of direct contact of the law school with the admitted students: the admission bulletin, direct mail marketing pieces, official law school forum, law school representative, and visits to the campus. The criterion set contained three

variables that were measurements of outside influences, or influences on the admitted students that are beyond the control of the law school: undergraduate pre-law advisor, attorneys, and parental influences.

According the Barcikowski and Steven's (1975) Monte Carlo study on the stability of the canonical coefficients and correlations, the number of subjects per variable required to achieve reliable results in interpreting the largest canonical correlation should be at least 20/1. When considering the two largest canonical correlations, a ratio ranging from 42/1 to 68/1 should be considered. In this example, the ratio of 31/1 used in this analysis is sufficient to achieve stable and reliable coefficients.

Analysis Results

To obtain an understanding of how well the variables are related to one another, Pearson correlations were calculated among each pair of variables, both within and across sets (see Table 1). The degree to which two predictors correlate is the degree to which they are said to be collinear. The collinearities among the direct contact measurements revealed some moderate values--the largest between FORUM and REPRESENTATIVE, and between ATTORNEY and PARENTS. When considering the between correlations, one notices that ADVISOR is moderately correlated with three of the five direct contact measures and that REPRESENTATIVE is appreciably correlated with each of the outside influence measures.

In canonical correlation analysis, an important measure to consult is the canonical correlation coefficient ($R_c$). According to Thompson (1984), conventional canonical correlation analysis initially begins by "collapsing each person's scores on the variables in

each variable set into a single composite variable." The bivariate correlation between the composite scores for the two sets of variables is the canonical correlation. As explained by Tatsuoka (1971), the total number of possible canonical correlations is equal to min($p$,$q$) where $p$ is the number of variables in the first set and $q$ is the number of variables in the second set. Therefore, in this example, there are three [min(5,3) = 3] canonical correlations yielded by the analysis.

As seen in Table 2, the two largest canonical correlations, $Rc_1$=0.65 and $Rc_2$=0.27, are both statistically significant at the 0.05 level. One also notices that $R_1$ is larger than any of the between-set correlations. According to Stevens (1996), even though a canonical correlation can be found to be statistically significant, a weak canonical correlation ($R_c$<0.30, $R_c^2$<0.09) may be trivial and of little practical value. Therefore, the researcher may decide a trivial canonical function is not worth interpreting. Because there is such a large decrease in value between $R_1$ and $R_2$ and because $R_2^2$=0.07 is small, only the first canonical function will be interpreted. Results from all three functions are presented in Table 3.

Result Interpretation

Once a canonical function is identified for interpretation, a number of coefficients may be calculated and consulted to answer various research questions (Thompson, 1984). Of interest to researchers is the contribution a single variable makes to the explanatory power of the set of variables to which the variable belongs. Two coefficients that address this question are the canonical function and structure coefficients. Similar to beta weights in regression, standardized function coefficients are weights applied to the

standardized data, which is summed to create the synthetic variables, or canonical variates (Thompson, 1991). When observing the standardized function coefficients for the direct contact measurements in the first function, one notices that FORUM and REPRESENTATIVE appear to be making the largest contribution, with the other three variables making contributions that are small and similar in size. For the outside influence variables, ADVISOR is making over twice the contribution as either of the other two variables.

As Kerlinger & Pedhazur (1973), Levine (1977), and Meredith (1964) argue, it is important to interpret canonical results based on not only function coefficients, but on structure coefficients as well. Structure coefficients are the bivariate correlations between the predictor variables and the synthetic variable created by the linear combinations, and generally take into account the collinearity, or overlap, of the set of variables. In this example, function and structure coefficients yield similar results. When observing the standardized structure coefficients for the direct contact measurements in the first function, FORUM and REPRESENTATIVE are making the largest contribution, with the other three variables making contributions that are smaller and similar in size. For the outside influence variables, ADVISOR is making the largest contribution. To obtain an estimation of the proportion of variance a variable shares with its canonical composite, the structure coefficient is squared. According to Table 3, FORUM and REPRESENTATIVE account for 80% and 74% of the direct contact variate, respectively, with BULLETIN, MAIL, and CAMPUS accounting for much smaller proportions of the variate. For the outside influence variable set, ADVISOR accounts for 78% of the variate, while ATTORNEY and PARENTS each account for less than 40%.

By summing the squared structure coefficients either across the functions or across the variables within a given function, one obtains the next two coefficients of interest: communality and adequacy. The communality coefficient for a variable (represented by $h^2$) equals the sum of the squared structure coefficients for all the functions and is an indication of what proportion of the variable's variance is reproducible. In other words, how useful the variable was in defining the canonical solution (Thompson, 1984). As seen in Table 3, the communality coefficients indicate that the researcher is not getting as much from the BULLETIN and MAIL variables as from the FORUM, REPRESENTATIVE, and CAMPUS variables.

The adequacy coefficient for a given function is the average of the squared structure coefficients for all the variables in the set and indicates how adequately the analysis represents the total variance in the unweighted set. In this example, the first function has a much larger adequacy coefficient than the other two functions. although the difference is more sizable for the set of variables measuring the direct contact methods.

Also of interest to the researcher is the relationship between the individual variables in one variable set with the canonical variates in the other variable set. In other words, what is the extent to which a variable contributes to explaining the composite, or linear combination of the variable set to which the variable of interest does not belong? The coefficient that addresses this question is referred to as an index coefficient. An index coefficient is the correlation between an unweighted variable in one set and the weighted and aggregated variables in the other set (Thompson, 1984). As seen in Table 3, ADVISOR has the largest index coefficient in the set of direct contact measurements,

and FORUM and REPRESENTATIVE have the largest index coefficients in the set of outside influence measurements.

The final component of canonical correlation analysis is the computation of redundancy coefficients. For a variable set on a function, a redundancy coefficient (*Rd*) is computed by multiplying the adequacy coefficient for the set by $R_c^2$ for the function. It indicates the average proportion of variance for variables in one set that is reproducible with (e.g., redundant with) the variables in the other set. Table 3 shows the *Rd* for each function.

It is often argued that redundancy coefficients should only be interpreted in the "few concurrent validity applications in which both variable sets consist of the same variables" (Thompson, 1991, p.89). Cramer and Nicewander (1979) argued that redundancy coefficients are not truly multivariate "in the strict sense because it is unaffected by the intercorrelaions of the variables being predicted. The redundancy index is only multivariate in the sense that it involves several criterion variables." (p. 43) Therefore, for the heuristic purposes of this paper, *Rd* values were computed and presented; however no interpretations or conclusions will be drawn considering that the research situation from which the data were drawn does not fit the application of redundancy coefficients suggested by Thompson (1991).

Summary

Canonical correlation analysis is a "parsimonious way of breaking down the association between two sets of variables through the use of linear combinations" (Stevens, 1986). As a result of the analysis, many types of coefficients can be generated

and interpretted. These coefficients are only considered stable and reliable if the number of subjects per variable is sufficiently large.

The first of these coefficients, the canonical correlation, is the bivariate correlation between the composite scores for the two sets of variables. Two additional coefficients, the canonical function and structure coefficients, address the contribution a single variable makes to the explanatory power of the set of variables to which the variable belongs. The communality coefficient explains how useful the variable is in defining the canonical solution. The adequacy coefficient indicates how adequately the analysis represents the total variance in the unweighted set. The extent to which a variable contributes to explaining the composite of the variable set to which the variable of interest does not belong is the index coefficient. A final outcome from canonical correlation analysis is the redundancy coefficient, which indicates the average proportion of variance for variables in one set that is reproducible with the variables in the other set. While the coefficient is easy to calculate, it is not recommended for interpretation in most cases.

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## TABLE 1

### Correlations Among the Direct Contact Measures

| | Bulletin | Mail | Forum | Representative | Campus |
|---|---|---|---|---|---|
| **Bulletin** | 1.0000 | | | | |
| **Mail** | 0.4630 | 1.0000 | | | |
| **Forum** | 0.3175 | 0.3902 | 1.0000 | | |
| **Representative** | 0.2294 | 0.3326 | 0.6370 | 1.0000 | |
| **Campus** | 0.2615 | 0.1923 | 0.1900 | 0.3128 | 1.0000 |

### Correlations Among the Outside Influence Measures

| | Advisor | Attorney | Parents |
|---|---|---|---|
| **Advisor** | 1.0000 | | |
| **Attorney** | 0.2267 | 1.0000 | |
| **Parents** | 0.2666 | 0.5140 | 1.0000 |

### Correlations Between the Direct Contact and Outside Influence Measures

| | Advisor | Attorney | Parents |
|---|---|---|---|
| **Bulletin** | 0.2896 | 0.1646 | 0.1831 |
| **Mail** | 0.3401 | 0.1663 | 0.1791 |
| **Forum** | 0.5466 | 0.3117 | 0.2858 |
| **Representative** | 0.4598 | 0.3946 | 0.3502 |
| **Campus** | 0.1350 | 0.2817 | 0.3018 |

## TABLE 2

### Canonical Correlations

| Function | Rc | $Rc^2$ | Pr>F |
|---|---|---|---|
| 1 | 0.6454 | 0.4165 | 0.0001 |
| 2 | 0.2723 | 0.0741 | 0.0112 |
| 3 | 0.0592 | 0.0035 | 0.8347 |

## TABLE 3

| Variable/ | Function I | | | | Function II | | | | Function III | | | | |
|---|---|---|---|---|---|---|---|---|---|---|---|---|---|
| Coefficient | Function | Structure | Structure2 | Index | Function | Structure | Structure2 | Index | Function | Structure | Structure2 | Index | $h^2$ |
| **Bulletin** | 0.1407 | 0.4836 | 0.2339 | 0.3121 | -0.0968 | -0.0861 | 0.0074 | -0.0235 | 0.3470 | 0.5409 | 0.2926 | 0.0320 | 0.5339 |
| **Mail** | 0.1131 | 0.5421 | 0.2939 | 0.3799 | -0.2560 | -0.2260 | 0.0511 | -0.0616 | 0.4638 | 0.4878 | 0.2379 | 0.0289 | 0.5829 |
| **Forum** | 0.5121 | 0.8944 | 0.8000 | 0.5773 | -0.6108 | -0.2900 | 0.0841 | -0.0790 | 0.1835 | -0.0133 | 0.0002 | -0.0008 | 0.8842 |
| **Representative** | 0.4263 | 0.8585 | 0.7370 | 0.5541 | 0.4606 | 0.2440 | 0.0595 | 0.0611 | -0.9154 | -0.4080 | 0.1665 | -0.0242 | 0.9630 |
| **Campus** | 0.1153 | 0.4045 | 0.1636 | 0.2610 | 0.8319 | 0.7854 | 0.6169 | 0.2139 | 0.5008 | 0.4293 | 0.1843 | 0.0254 | 0.9648 |
| **Adequacy** | | | 0.4457 | | | | 0.1638 | | | | 0.1763 | | |
| **Rd** | | | 0.1856 | | | | 0.0120 | | | | 0.0006 | | |
| $Rc^2$ | | | 0.4165 | | | | 0.0741 | | | | 0.0035 | | |
| **Advisor** | 0.7509 | 0.8842 | 0.7818 | 0.5707 | -0.7232 | -0.4586 | 0.2103 | -0.1249 | 0.0496 | 0.0885 | 0.0078 | 0.0052 | 1.0000 |
| **Attorney** | 0.3481 | 0.6233 | 0.3885 | 0.4023 | 0.4754 | 0.6139 | 0.3769 | 0.1672 | -1.0140 | -0.4844 | 0.2346 | -0.0287 | 1.0000 |
| **Parents** | 0.2042 | 0.5833 | 0.3402 | 0.3765 | 0.5883 | 0.6399 | 0.4095 | 0.1743 | 1.0083 | 0.5003 | 0.2503 | 0.0296 | 1.0000 |
| **Adequacy** | | | 0.4264 | | | | 0.2078 | | | | 0.1122 | | |
| **Rd** | | | 0.2098 | | | | 0.0246 | | | | 0.0006 | | |



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